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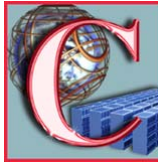
# **Parallel Graph Algorithms for Complex Networks**

**Edmond Chow,  
Tina Eliassi-Rad, Keith Henderson**

***Lawrence Livermore National Laboratory  
Comp/NAI LDRD Project***

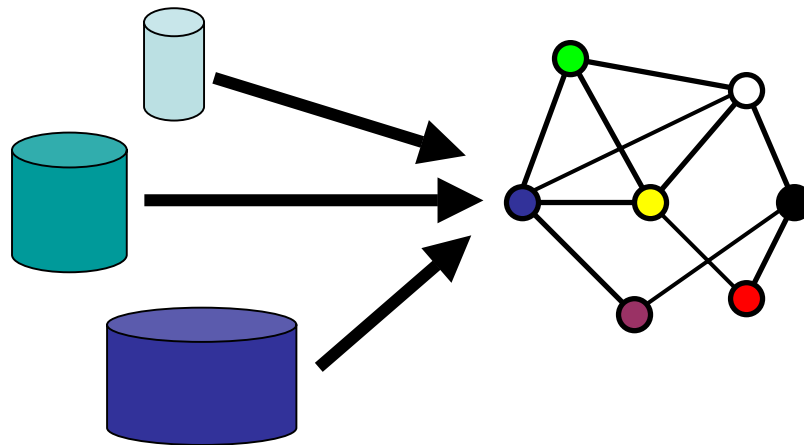
UCRL-PRES-203191

**DHS Advanced Scientific Computing PI Meeting  
March 16, 2004**



# Intelligence analysts must find relationships in huge amounts of data

- Data is collected from multiple sources at increasing rates
- Challenge: identify relationships and uncover patterns in a timely manner
- Approach: use *semantic graphs* to represent the data and graph algorithms to discover hidden relationships

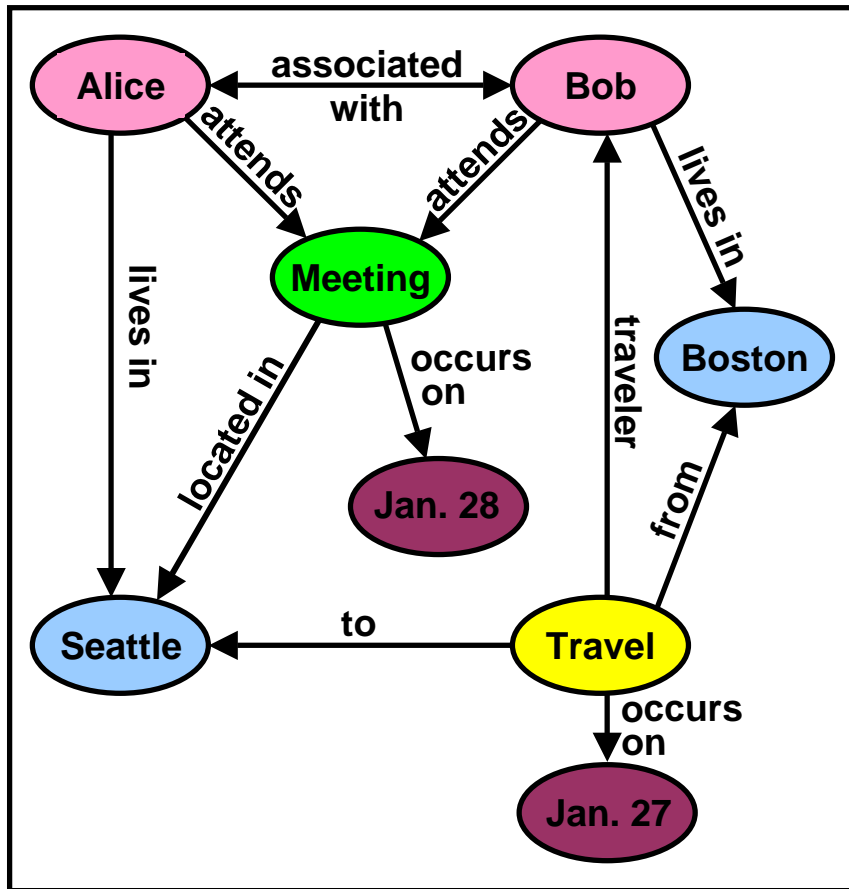




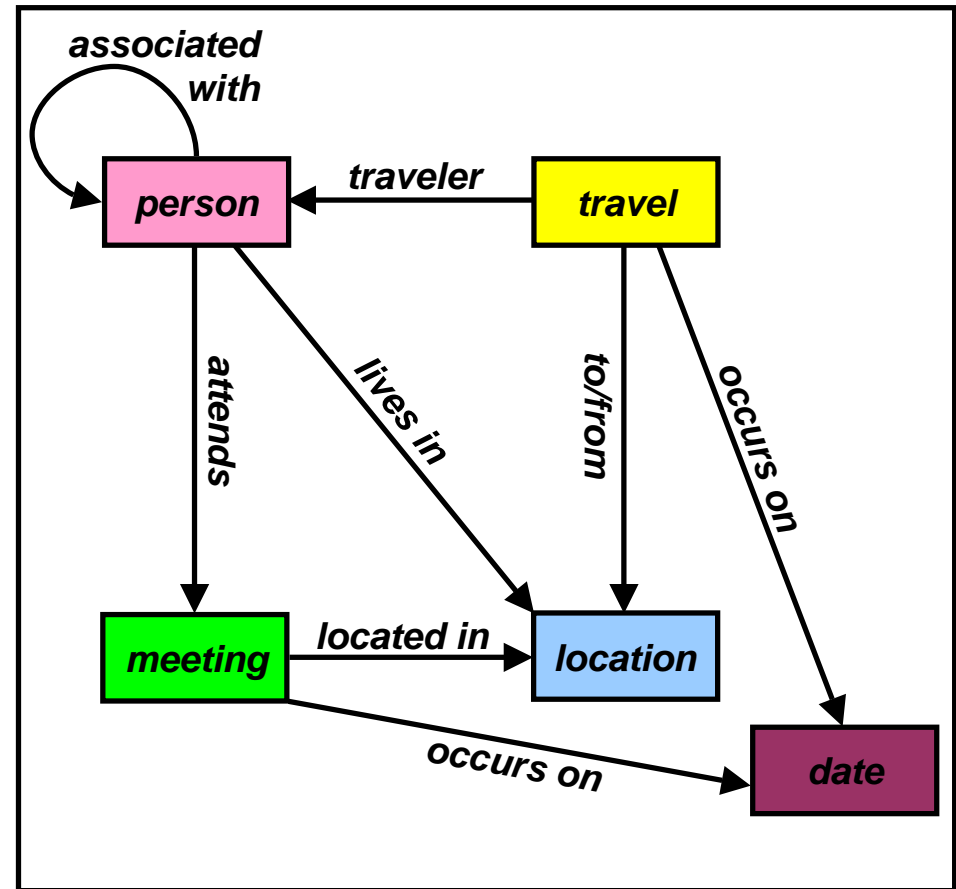
# Semantic graphs have types and attributes on the vertices and edges



Semantic Graph



Ontology





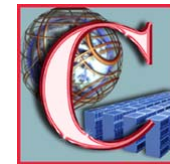
# **Semantic graphs in intelligence applications are becoming enormous**

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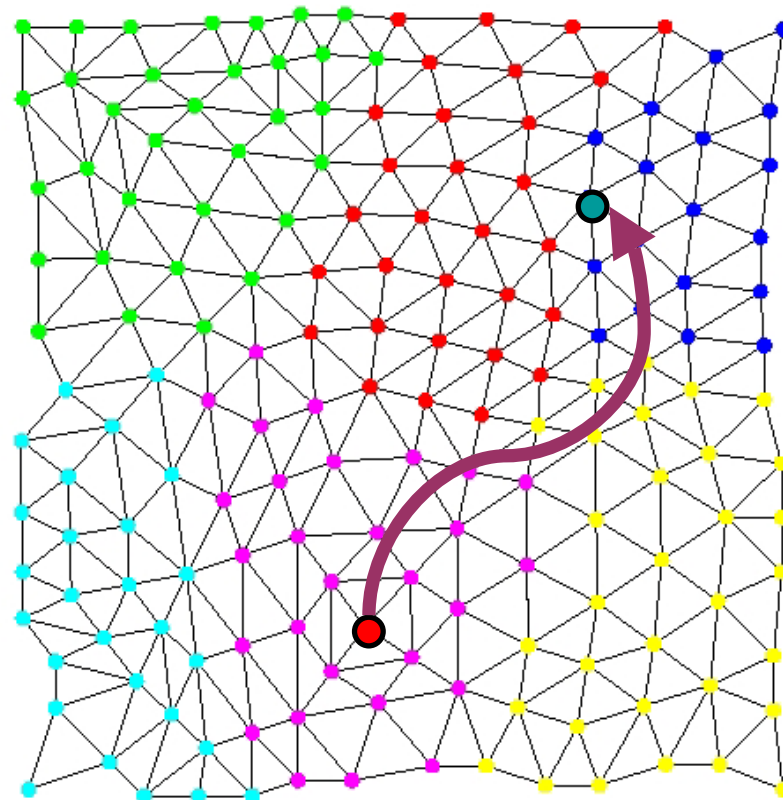
- **Distributed memory parallel computers must be used to store and search these graphs**
- **Graphs must be partitioned onto separate memories and graph searches must have low communication cost**
- **Semantic graphs have topological properties that make partitioning very challenging**
- **Our goals are to develop partitioners and efficient parallel search algorithms**



# We are developing parallel algorithms to search massive graphs

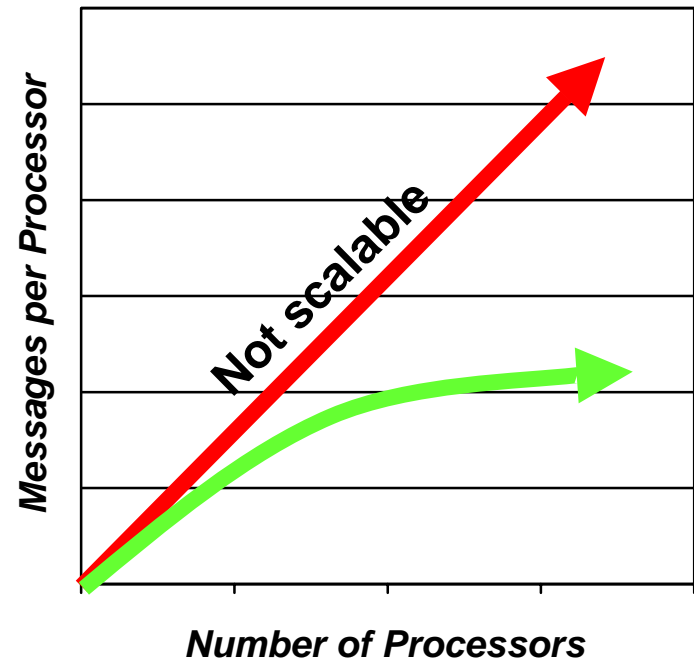
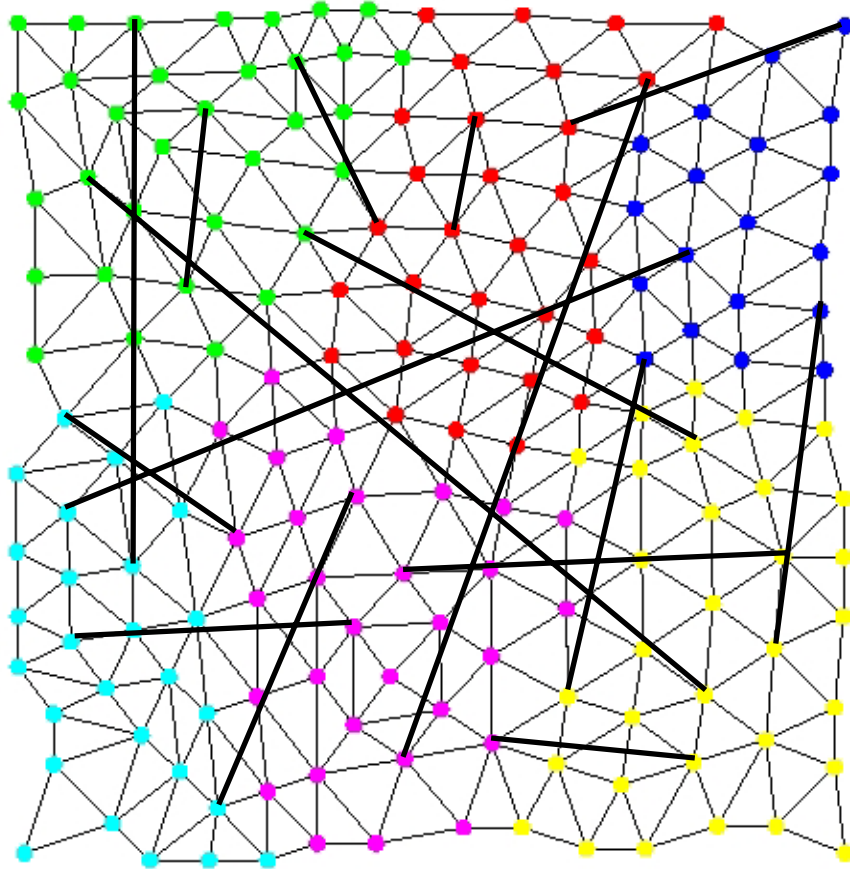


- Semantic graph models
- Partitioning approaches with bounded messaging cost that utilize the ontology
- New ontology-based probabilistic heuristics for searching graphs
- New heuristics for searching general graphs



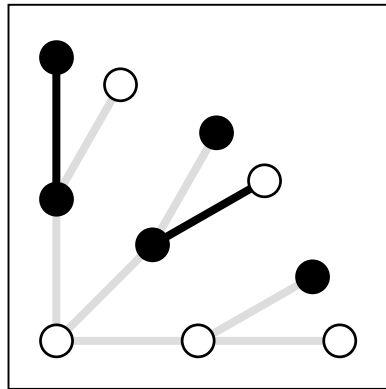
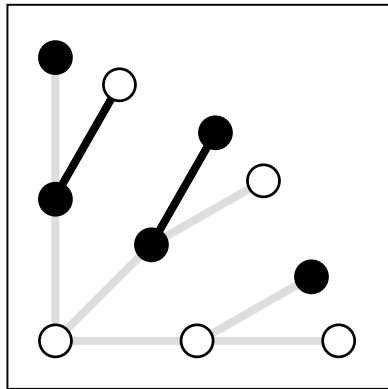
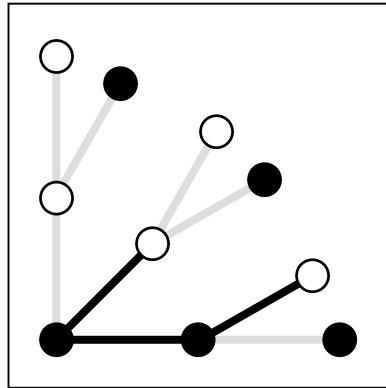
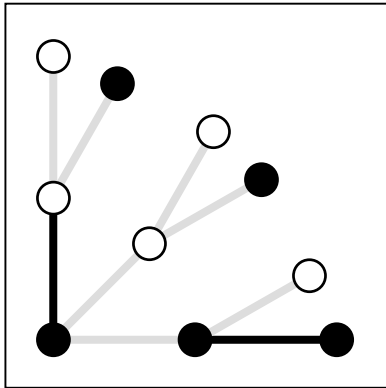


# Partitions of a complex network may induce $O(P^2)$ communication





# We use an edge partitioning approach to get $O(P)$ communication



Example partitioning on 4 processors

- Coarse partitioning of  $\sqrt{P}$  parts
- Expanding a vertex is  $\sqrt{P}$  communication
- Updating the fringe is also  $\sqrt{P}$  communication
- Disadvantage is more complicated parallel implementation
- Edge partitioning can be based on edge types



# 2D-hypergraph partitioning shows dramatic potential of this approach



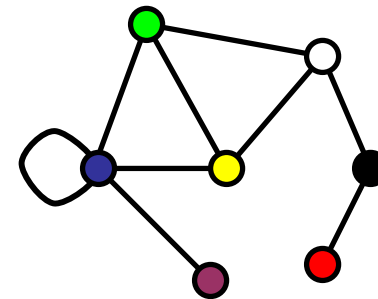
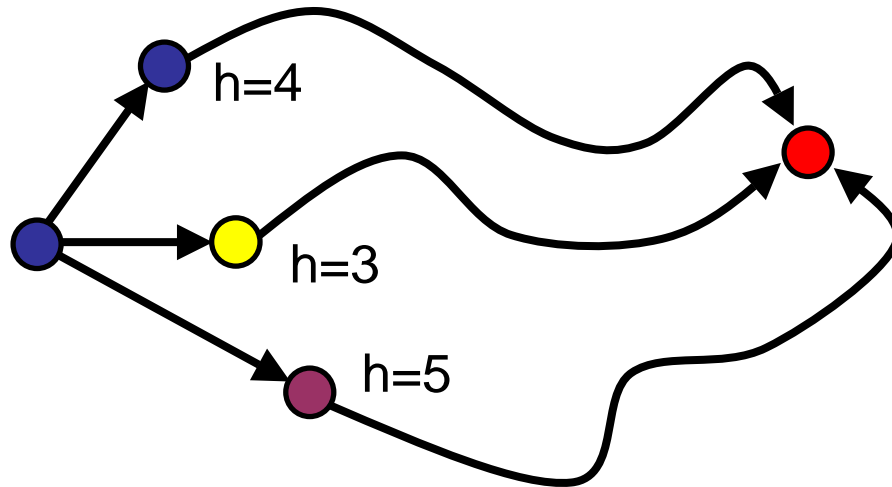
	<i>Vertices</i>	<i>Edges</i>	<i>Num. Messages</i>		<i>Comm. Volume</i>	
			<i>1D</i>	<i>2D</i>	<i>1D</i>	<i>2D</i>
<b><i>Mesh</i></b>	205761	615360	2000	160	2136	2296
<b><i>Random</i></b>	205761	615360	157200	760	158668	159576
<b><i>Scale-Free</i></b>	205761	617277	156800	760	201786	221174
<b><i>Spatial</i></b>	51682	154080	22400	760	23126	23922
<b><i>Web</i></b>	325729	1090108	20800	680	31727	44246

**Simulated communication performance for 400 partitions shows potential for orders of magnitude improvement with this approach**





# We are developing heuristics to guide and accelerate the search



The ontology gives a lower bound on distance

- Distance estimates are used to guide the heuristic search (A\* search)
- Vertex and edge frequencies give a probabilistic measure of the heuristic estimate
- Parallelization will use standard owner-computes model



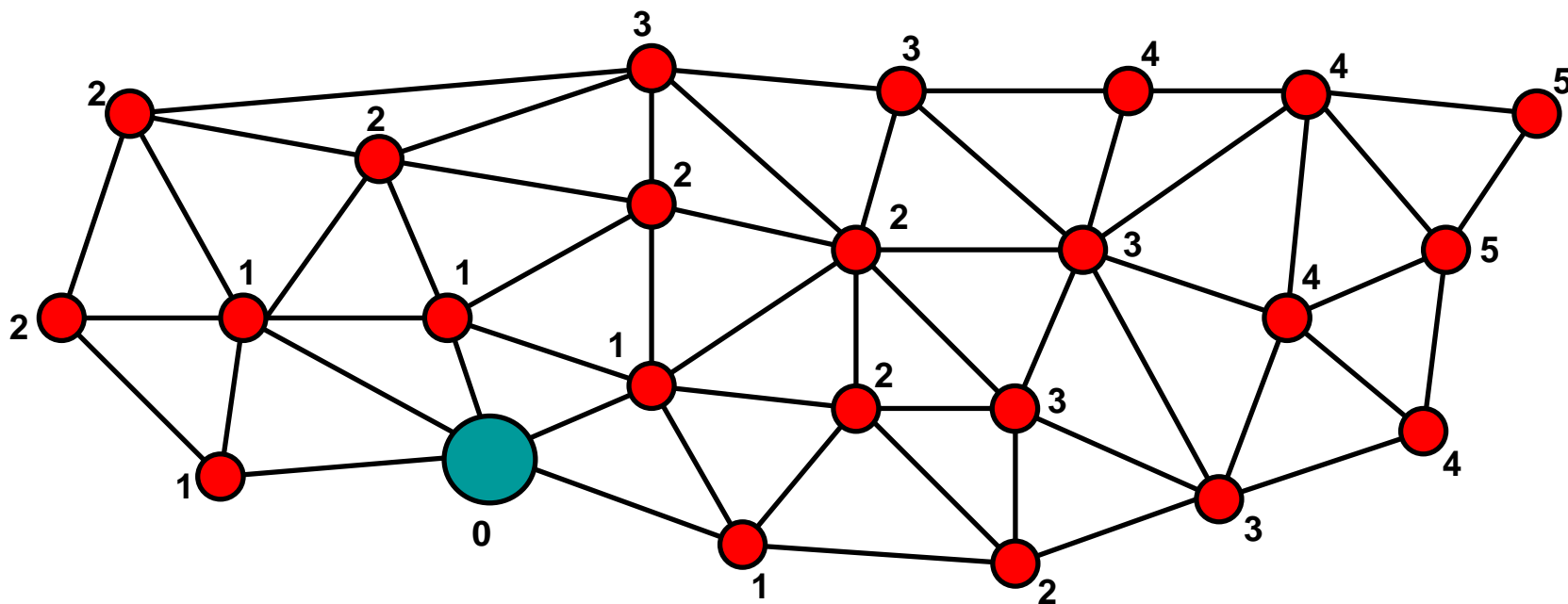
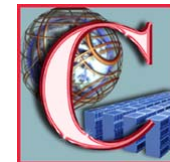
# For typical ontologies, heuristic search is 2-3 times faster than BFS



<i><b>Ontology edge density</b></i>	<i><b>BFS vertices visited</b></i>	<i><b>Ontology heuristic work ratio</b></i>	<i><b>Probab. heuristic work ratio</b></i>
0.05	26594	0.483	0.323
0.25	26443	0.655	0.495
0.50	24533	0.798	0.634
0.75	22534	0.885	0.676
1.00	23924	1.000	0.779



# We are also developing heuristics based on the graph itself

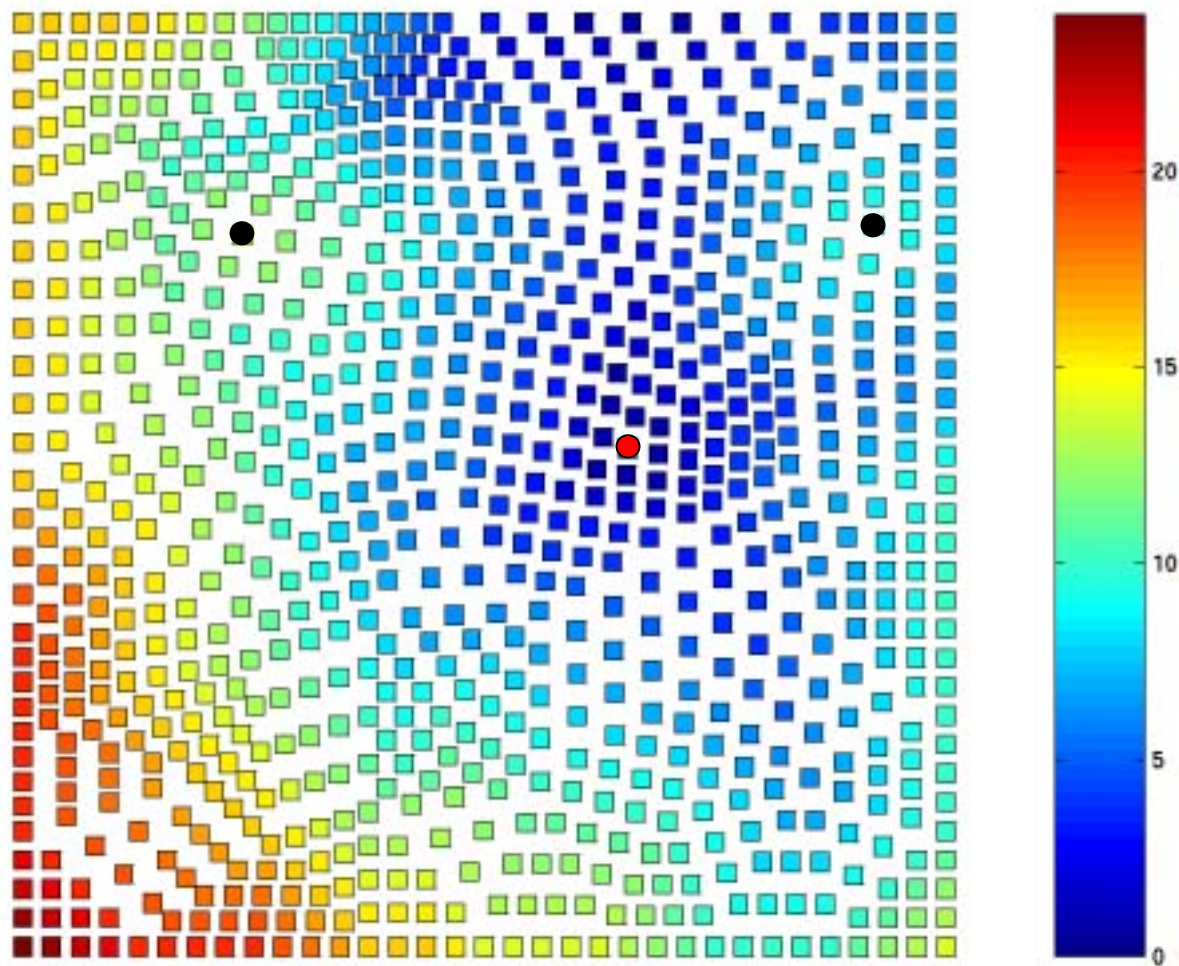


$$h_1(i,j) = | \text{level}(i) - \text{level}(j) |$$

$$h(i,j) = \max\{ h_1(i,j), \dots, h_m(i,j) \}, \quad m \text{ centers}$$



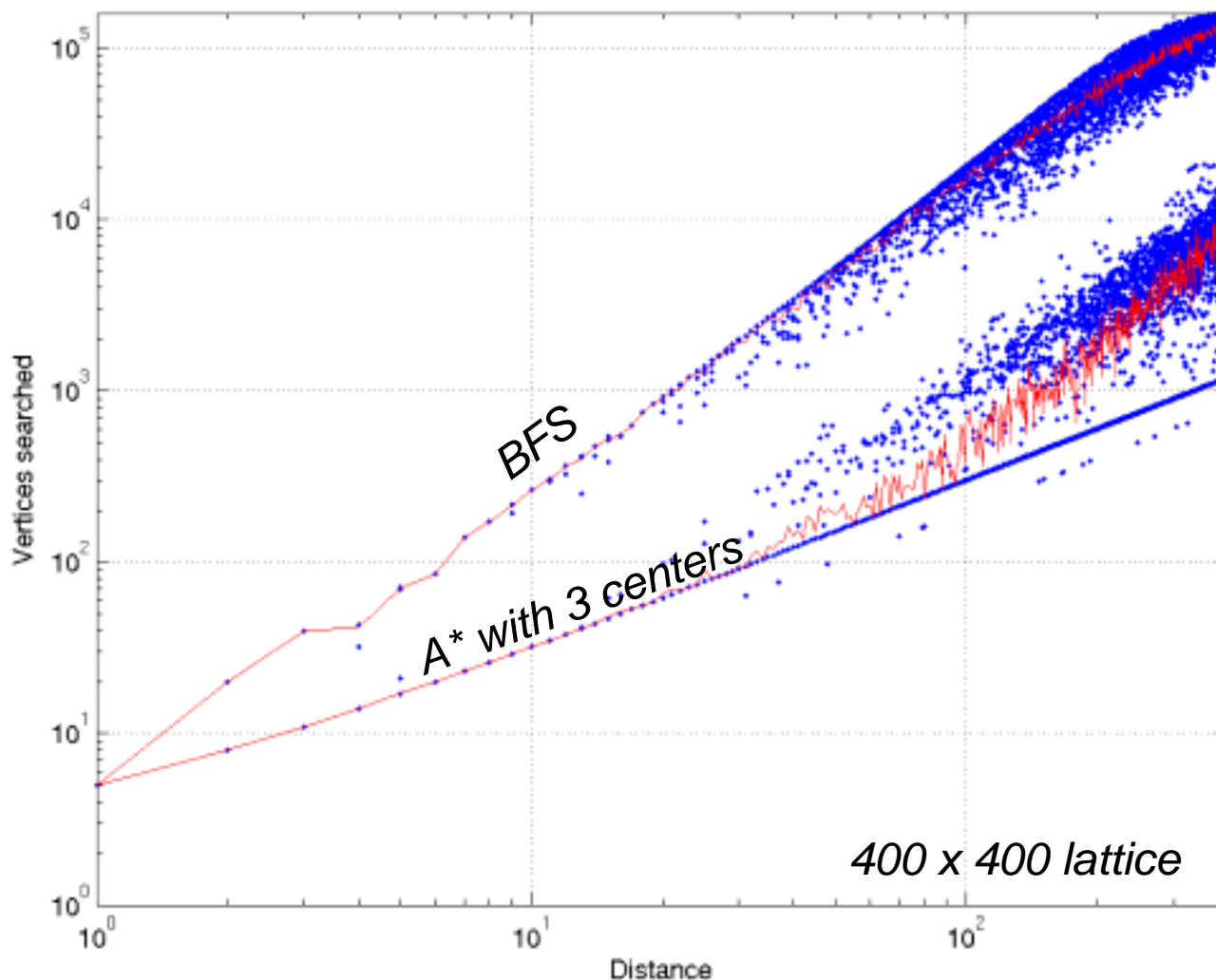
# Example: heuristic distance to a vertex on a mesh



Heuristic distance to red vertex given two center vertices

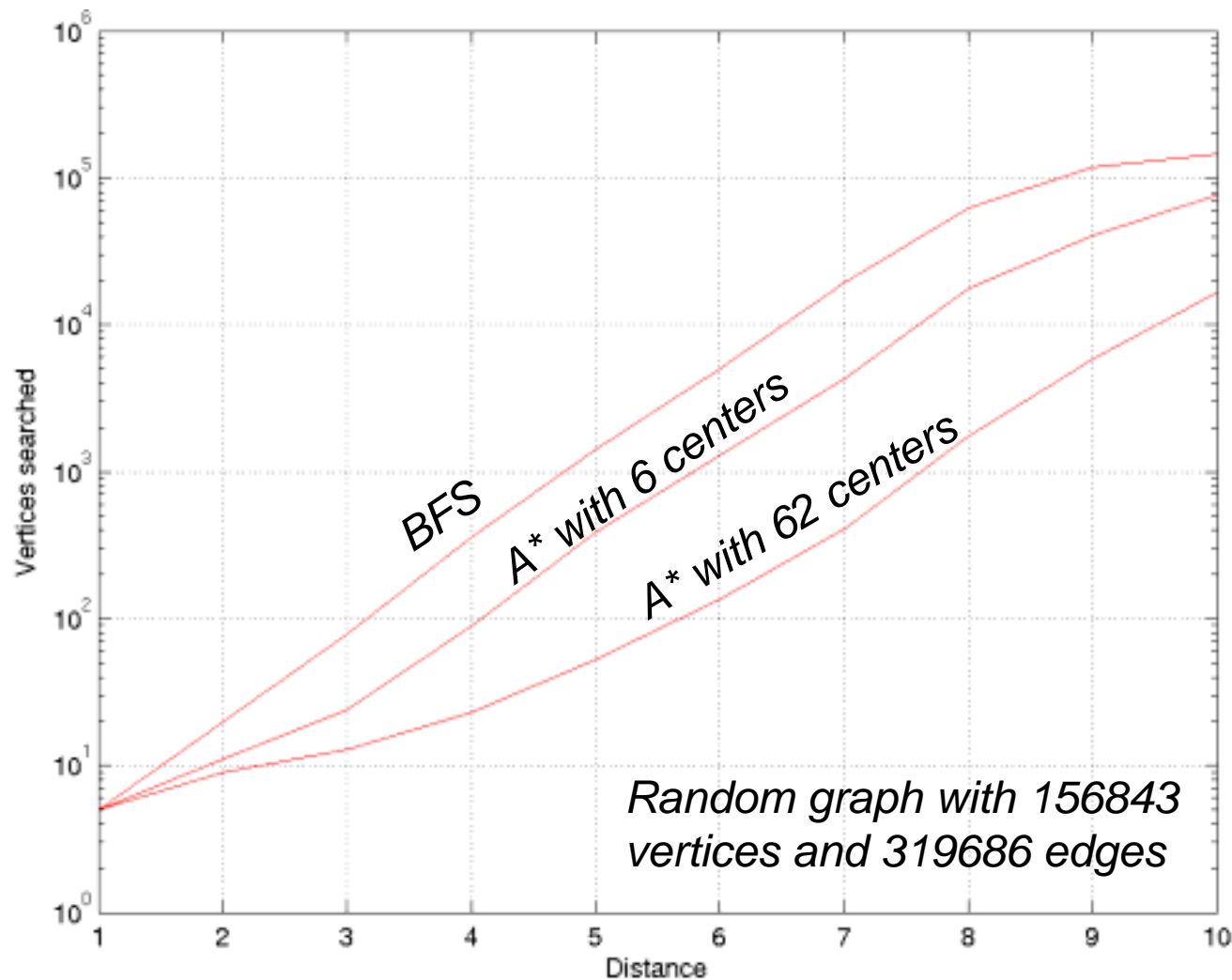


# For a lattice graph, search complexity reduces from $O(L^2)$ to $O(L)$





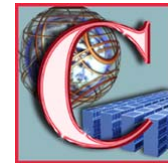
# For a random graph, search time can be reduced ten-fold





# **Our new algorithms will give graph analysis vastly more power**

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- **Enable search on distributed parallel computers**
- **Anticipate significant decreases in communication time and search complexity**
- **Future research will continue to be relevant to large-scale applications**
  - **Develop parallel implementations**
  - **Exploit properties of semantic graphs and complex networks in 2D partitioning**
  - **Develop hierarchical graph representations and algorithms**
  - **Develop algorithms for dynamically changing graphs**
  - **Exploit temporal locality in queries to drive partitioning**



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**This work was performed under the auspices of the  
U.S. Department of Energy by University of California Lawrence  
Livermore National Laboratory under contract no. W-7405-Eng-48.**